ASSIGNMENT - 2

1) p=0.25 n=1000

The simulation has been done in the file Q1.py

Jo get the result, we this command in Terminal:

python3 Q1.py

The output comes out to be:

Probability of at least 240 A's in the sequence = 0.7784

Explanation of code 8

for each simulation we generate a random sequence of n=1000 numbers, as follows:

- i) generate a random number from 0 to 1
- in our case) then add a 1 to the sequence, otherwise of the sequence.

Counting the number of times we get a sum of 240 or more and then dividing it by the number of simulations, gives the answer.

Normal Approximation to Binomial Distribution $p = 0.25, q = 1 - p = 0.75, n = 1000, \mu = n \times p = 250$

By central limit theorem, we can approximate the binomial distribution into a normal distribution. On com

5 = Inpq = J1000×0.25×0.75 = 13.69

 $Z = \frac{240 - \mu}{6} = \frac{240 - 250}{13.69} = -0.7305$

On comparison, we see both the values are very close.

Using z-score table, we get z score corresponding to 0.7305 as $P(x \ge 240) = 1 - P(x < 240) = 1 - 0.2327 = 0.7673$

$$P = 0.3 \quad n = 10 \qquad P(x = 0), P(x = 2), E(x), V_{an}(x)$$

$$P = 0.3 \quad q = 1 - P = 0.7 \quad n = 10$$

$$P(x = x) = {}^{10}C_{x}P_{q}^{x}$$

$$P(x = 0) = {}^{10}C_{0}0.3^{\circ}0.7^{\circ} = 0.02824$$

$$P(x = 2) = {}^{10}C_{2}0.3^{\circ}0.7^{\circ} = \frac{10x_{q}}{2} \times 0.3^{2} \times 0.7^{\circ} = 0.23347$$

$$E(x) = n \times P = 10 \times 0.3 = 3$$

$$Van(x) = 2n \times P \times q = 10 \times 0.3 \times 0.7 = 2.1$$

3) Applications of k-mer analyses &

K-mers with $k \ge 2$ are used to identify regions with aberrant base compositions that show genome segments obtained by lateral transfer.

The different applications are as follows &

different codon usage frequencies that differs from between organisms. Hence, they can be used to identify horizontally transferred genes.

Observed prequencies of k-words can be used to analyze DNA sequences.

iii) k-tuple (k > 3) prequencies may also consist of be useful in predicting whether an unannotated sequence is coding or each-coding.

strains/species. Therefore, bacterial genomes can be clustered into natural groups on the basts of k-mer distribution similarities.

4) The two sequences are:

GGCTGCAACTAGCTC

GGGTAAGCTTGC

	G	G	G	T	A	A	G	C	-	Г	T	9	(
G	X	X	×				X					X	
G	X	X	X				X					X	
C								X					X
T			1	X					X)	<		
G	X	×	×				X				1	X	
C								X			-		X
A					X	X							^
A					X	X			3		+	+	
C								X	-	-	-	-	X
T				X				/\	X	X	+	+	_
A					X	X			/\	^		+	
G	X	X	X				X				X		
C								X				-	<
T				X					X	X			
C								X				>	1

The colored crosses denote a match. The conserved regions of length > 2 are:

i) AGCT (sequence 1:11 to 14, sequence 2:6 to 9 (included)): in violet

ii) GCT (sequence 1:2 to 7, sequence 2:7 to 9 (included): in orange

(i) TGC sequence 1:4 to 6, sequence 2:9 to 11 (included): in green

Pseudocode for Q1:

specify pong no. of simulations

count of 240 A's initialize to O

for a > 0 to (no. of simulations -1):

X = random array of length in

If any element in X <p, then X (of that index) = 1.

Add all elements.

If (sum > 240) then - count of 240 A's +=1

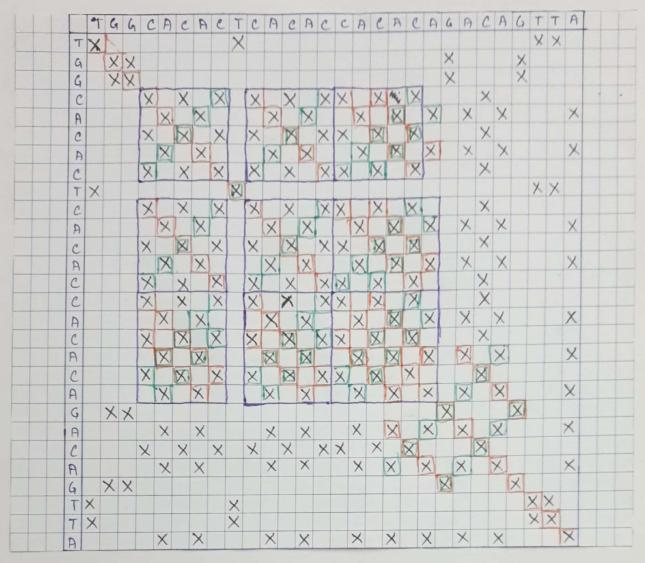
(Heter for 100b)

Probability =

count of 240A's

no of simulations

5) Identifying repeat region: TGGCACACTCACACCACACAGACAGTTA



Mainly two kinds of repeats:

- 1) Low complexity regions in violet

 They can be identified by horizontal/vertical rows of dots/that can merge into reetangular or square patterns
- 2) Internal repeat regions
 - (i) Forward repeats in orange

 They can be identified as crosses(X) or dots(.) trying to

 torm a torward diagonal (hop left bottom right)
 - (ii) Inverted repeats in green

 They can be identified as crosses(X) or dots() trying to form a backward diagonal (bottom-left hop right)

Pseudocode for QJ: matches marked as "x" show-# The code gives a string or def deplot (seg1, seg2): # the whole table as it should dotplot = "1 1" for j > 0 to length(L2)-1: # be shown dotplot += L2[i]+" # L1, L2 are lists formed from for j-> 0 to length(L2)-18 # seq1, seq2 respectively as dotplot += "In1" # they were strings dotplot += '-1' dotplot += '-1/p' for i- 0 to length (L1)-1: of Checking matches dotplot += '1'+11+11' # fore each elevent for j -> 0 to length (L2)-1: val = "x" y (L1[i] = L2[j]) else " " # of both (ls) # sequences dotplot += val + 1 dotplot += 'n' return dotplot Function name is show-plot which takes two string The news the consts as input Main modules Seg = "TGGCACACTCACACACACACAGACAGTTA" # given in question print (desplot (seq, seq)) # since self-matching.

Scanned with CamScanner

First requerce is lated in 5' to 3' direction in the horizontal direction and its complementary requerce is lated along the vertical direction also in the 5' to 3' direction.

Matrix then tracks the identical matches.

Self-complementary megions can be seen as diagonals from top left to lower right

	A	U	6	U	G	6	C	A	U	6	C	C	A	G	G
C							X				X	X			
CCU							X				X	X			
U		X		X					X						
9			X		X	X				X				X	X
G			X		X	X				X				X	X
C							X				X	X			
A	X							X					X		
U		X		X					X	1					
G			X		X	X				X				X	X
			al .				X		7.		X	X			
C							X				X	X			
A	X							X					X		
C							X				X	X			
A	X							X					X		
U		X		X					X						

Sequence:
AUGUGGCAUGCEAGG

Complementary Sequence:
CCUGGCAUGCCACAU

Longest self-complementary regions found here is shown in VIOLET (length = 10) and is: UGGCAUGCCA

Other small regions marked in green: AUG, CAU

Dotplots are also shown in the form of code for questions 4,5,6. The codes (files in py') are Q4.py, Q5.py and Q6.py respectively.